CLAIMS

| 1 | 1. A crystallization parameter optimization process comprising the |
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| 2 | steps of: |
| 3 | selecting a plurality of physical characterization input variables to |
| 4 | define a total crystallization experiment permutation number for a crystallant; |
| 5 | performing a plurality of crystallization experimental samples, said |
| 6 | plurality of crystallization experimental samples being less than the total |
| 7 | crystallization experiment permutation number; |
| 8 | training a predictive crystallization function through analysis of said |
| 9 | plurality of crystallization experimental samples; and |
| 10 | determining an optimal physical crystallization parameter from said |
| 11 | predictive crystallization function. |
| | |
| 1 | 2. The process of claim 1 wherein said predictive crystallization |
| 2 | function is a neural network. |
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| 1 | 3. The process of claim 1 wherein said crystallant is a protein. |
| | |
| 1 | 4. The process of claim 1 wherein each of said plurality of physical |
| 2 | crystallization input variables is selected from a group consisting of: |
| 3 | temperature, protein dilution, anionic precipitate, organic precipitate, buffer |
| 4 | pH, precipitation strength, organic moment, percent glycerol, additive, divalent |
| 5 | ion, gravity, light, magnetism, atmosphere identity, and atmosphere pressure. |
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| 1 | 5. The process of claim 1 wherein the plurality of crystallization |
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| 2 | experimental samples performed is less than 5% of the total crystallization |
| 3 | experiment permutation number. |
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| 1 | 6. The process of claim 1 wherein the plurality of crystallization |
| 2 | experimental samples performed is less than 0.1% of the total crystallization |
| 3 | experiment permutation number. |
| | |
| 1 | 7. The process of claim 1 wherein said predictive crystallization |
| 2 | function analyzes a crystallization experimental sample as to a status selected |
| 3 | from the group consisting of: clear drop, phase change, precipitate, and |
| 4 | spherulettes. |
| | |
| 1 | 8. The process of claim 1 wherein said predictive crystallization |
| 2 | function trains through back propagation. |
| | |
| 1 | 9. The process of claim 8 wherein said predictive crystallization |
| 2 | function includes a hidden layer intermediate between input values and said |
| 3 | optimal physical crystallization parameter. |
| | 2 11 |
| 1 | 10. The process of claim 1 wherein the performance of said |

plurality of crystallization experimental samples is automated.

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- 1 11. The process of claim 10 further comprising the step of communicating said plurality of physical crystallization input variables between a manufacturing execution system performing said plurality of experimental samples and said predictive crystallization function.
- 1 12. The process of claim 1 further comprising the step of communicating said predictive crystallization function to a database.
- 1 13. The process of claim 12 wherein said database includes 2 characteristics of a crystallization sample.
- 1 14. The process of claim 1 further comprising the steps of 2 attempting crystal growth using said optimal physical crystallization parameter.
 - 15. The process of claim 14 further comprising the step of communicating on said crystal growth attempt to a shared database.
 - 16. The process of claim 15 further comprising the step of classifying said crystal growth attempt on a basis selected from the group consisting of: said optimal physical crystallization parameter, said predictive crystallization function, and a physical property of a crystallant.

| 1 | 17. The process of claim 1 wherein performing said plurality of |
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| 2 | crystallization experimental samples comprises the steps of: |
| 3 | controlling a plurality of variables where each of said plurality of |
| 4 | variables assumes an index value or plurality of index values; and |
| 5 | performing a Chernov analysis to derive a minimized combined |
| 6 | quantity representative of said total crystallization permutation number. |
| 1 | 18. The process of claim 1 wherein said plurality of crystallization |
| 2 | experimental samples are converted to vectors prior to the training of said |
| 3 | predictive crystallization function. |
| 1 2 | 19. The process of claim 18 further comprising the step of clustering said vectors. |
| 1 | 20. The process of claim 19 wherein clustering occurs through the |
| 2 | application of an analysis selected from the group consisting of: a neural net, a |
| 3 | Chernov algorithm, a Bayesian net, a Bayesian classification schema, and a |
| 4 | Bayesian decomposition. |
| | |
| 1 | 21. A crystallization parameter optimization process comprising the |
| 2 | steps of: |
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| selecting a plurality of physical characterization input variables for a |
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| known crystallant to define a total crystallization experiment permutation |
| number; |
| performing a plurality of crystallization experimental samples on said |
| known crystallant; |
| training a predictive crystallization function through analysis of said |
| plurality of crystallization experimental samples; |
| determining an optimal physical crystallization parameter for said |
| known crystallant; |
| storing said optimal physical crystallization parameters and a physical |
| property of said known crystallant sample in a classification system; and |
| comparing an unknown crystallization sample to the classification of |
| said known crystallant. |
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| 22. The process of claim 21 wherein said predictive crystallization |
| function is a neural network. |
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| 23. The process of claim 21 wherein said classification system is |
| based on an aspect selected from the group consisting of: nodal basis functions, |
| nodal construction similarities, and contribution of a particular physical |
| characterization input variable. |
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| 2 | relates said known crystallant and said unknown crystallization sample. |
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| 1 | 25. The process of claim 21 wherein said classification system is |
| 2 | self-learning. |
| 1 2 | 26. The process of claim 21 wherein said classification system is self-organized. |
| | |
| 1 | 27. The process of claim 21 wherein each of said plurality of |
| 2 | physical crystallization input variables is selected from a group consisting of: |
| 3 | temperature, protein dilution, anionic precipitate, organic precipitate, buffer |
| 4 | pH, precipitation strength, organic moment, percent glycerol, additive, divalent |
| 5 | ion, gravity, light, magnetism, atmosphere identity, and atmosphere pressure. |
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| 1 | 28. The process of claim 21 wherein the performance of said |
| 2 | plurality of crystallization experiments is automated. |
| | |
| 1 | 29. The process of claim 21 further comprising the steps of |
| 2 | attempting crystal growth using said optimal physical crystallization parameter. |
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| 1 | 30. The process of claim 21 wherein performing said plurality of |
| 2 | crystallization experimental samples comprises the steps of: |
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The process of claim 21 wherein a comparative neural network

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| 3 | controlling a plurality of variables where each of said plurality of |
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| 4 | variables assumes an index value or plurality of index values; and |
| 5 | performing a Chernov analysis to derive a minimized combined |

6 quantity representative of said total crystallization permutation number.

- database wherein said shared database also stores at least one type of protein information selected from the group consisting of: protein expression gene; protein characteristics; protein class hierarchy; actual protein chemical structure including primary, secondary, tertiary and where applicable quaternary structures; protein crystal generation recipe parameters; and optimal crystallization screen design.
- 32. A protein crystal derived by the process of claim 1.
- 1 33. A neural network having been trained through analysis of a 2 plurality of crystallization experimental samples to predict optimal 3 crystallization conditions for a protein.
- 1 34. The network of claim 33 wherein said plurality of samples 2 comprises samples failing to yield crystals.

| 1 | 35. A system for crystallization parameter optimization, the system |
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| 2 | comprising: |
| 3 | a database having a plurality of input variables, each of said plurality of |
| 4 | input variables having a value range; |
| 5 | an incomplete factorial screen program having a trainable predictive |
| 6 | crystallization function; |
| 7 | a computer capable of executing the incomplete factorial screen |
| 8 | program to determine an optimal crystallization parameter; and |
| 9 | a manufacturing execution system for automatically acquiring of a |
| 10 | datum from each of a plurality of crystallization experimental samples, |
| 11 | analyzing and archiving of data from the incomplete factorial screen program. |
| | |
| 1 | 36. The system of claim 35 wherein said manufacturing execution |
| 2 | system controls at least one piece of crystallization hardware selected from the |
| 3 | group consisting of: a liquid dispenser, a crystallant dispenser, a robotic |
| 4 | handler, an imaging system, a sample centering motor relative to a camera |
| 5 | focal plane, and a lighting system. |
| | |
| 1 | 37. The system of claim 36 wherein said sample centering motor is |
| 2 | coupled to at least one of: a sample stage and said camera for automatically |
| 3 | positioning the specimen within the focal plane of said camera. |
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| 2 | indexing each of said plurality of samples. |
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| 1 | 39. The system of claim 36 further comprising a centering algorithm |
| 2 | coupled to said motor for converging a central region of the specimen with a |
| 3 | central region of the camera focal plane. |
| 1 | 40. The system of claim 39 wherein said centering algorithm operates automatically. |
| 1 | 41. The system of claim 35 further comprising a drop identification |
| 2 | algorithm for evaluating a liquid drop associated with each of said plurality of |
| 3 | samples. |
| 1 2 | 42. The system of claim 41 wherein the liquid drop is classified into a preselected plurality of classes. |
| 1 2 | 43. The system of claim 42 wherein said drop identification algorithm operates automatically. |
| 1 2 | 44. The system of claim 36 wherein said motor is coupled to said camera. |

The system of claim 35 further comprising a barcode for

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| 2 | interfaced with said robotic handler. |
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| 1 | 46. The system of claim 37 wherein said scheduling software is |
| 2 | interfaced with said sample stage. |
| 1 | 47. The system of claim 35 further comprising a database that stores |
| 2 | crystal relevant parameters. |
| 1 | 48. The system of claim 47 wherein said crystal relevant parameters |
| 2 | include at least one parameter of the group consisting of: crystal weight, |
| 3 | crystal specimen pH, crystal specimen temperature, crystal specimen protein |
| 4 | type, detergents present, additives present, preservatives present, reservoir |
| 5 | buffer present, reservoir buffer concentration, reservoir buffer pH, crystal |
| 6 | specimen volume, notes, crystal specimen score, and crystal specimen drop |
| 7 | descriptor. |
| 1 | 49. The system of claim 47 wherein said database is relational |
| 2 | between said predictive crystallization function and said crystal parameters. |
| 1 | 50. The system of claim 47 wherein said database is connected to a |
| 1 | · |
| 2 | structured query language database. |

The system of claim 36 further comprising scheduling software

- 1 51. A protein crystal derived from a system of claim 35.
- 1 52. A process according to claim 1 substantially as described herein
- 2 in any of the examples.